

Eigenvalues, square roots, ...

Knowing the conclusion of this page,
is not curriculum.

Being able to understand the calculations,
is curriculum. Take it as exercise.

Let \vec{V} have eigenvectors of \vec{A} as columns.

$$\begin{aligned} \text{Then } \vec{A} \vec{V} &= (\vec{A} \vec{v}^{(1)} \mid \dots \mid \vec{A} \vec{v}^{(n)}) \\ &= \vec{V} \vec{\Lambda} \text{ where } \vec{\Lambda} = \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_n \end{pmatrix} \end{aligned}$$

A symmetric matrix has n lin. indep eigenvectors,
Stacking these into \vec{V} , it is invertible.

$$\vec{A} = \vec{V} \vec{\Lambda} \vec{V}^{-1}$$

High powers? $\vec{A}^{2018} = \vec{V} \vec{\Lambda}^{2018} \vec{V}^{-1}$ $\ddot{\smile}$

Low powers, like $\dots \frac{1}{2}$? $\vec{V} \begin{pmatrix} \sqrt{\lambda_1} & 0 \\ 0 & \dots \sqrt{\lambda_n} \end{pmatrix} \vec{V}^{-1}$?

Ok if \vec{A} pos. semidef or pos. def.

Fact: a pos. semidef \vec{A} has a unique pos. semidef square root \vec{B} , often denoted $\vec{A}^{1/2}$.

If $\vec{Y} = \text{random } \sim E \vec{Y} = \vec{0}$, $E \vec{Y} \vec{Y}^T = \vec{A}$ invertible,

what is the covar of $\vec{Z} = (\vec{A}^{-1})^{1/2} \vec{Y}$?

Quadratic forms under linear constraints

Consider a restriction to a subspace: $\vec{B}\vec{x} = \vec{0}$
 (\vec{B} can be assumed to have full rank).

$Q = \vec{x}^T A \vec{x}$ is called pos. def. subject to $\vec{B}\vec{x} = \vec{0}$
 if $Q(\vec{x}) > 0$ for all $\vec{x} \neq \vec{0}$ such that $\vec{B}\vec{x} = \vec{0}$.

Pos semidef subject to $\vec{B}\vec{x} = \vec{0}$ if ...

Neg	_____	_____	...
Neg. def	_____	_____	...
Indef	_____	_____	...

Fact: \vec{A} pos. def \Rightarrow pos. def subject to $\vec{B}\vec{x} = \vec{0}$
but not \Leftarrow

⋮
 neg

neg

But Q indefinite subject to $\vec{B}\vec{x} = \vec{0} \Rightarrow \vec{A}$ indef.

Application: approximate a Lagrange problem

Criteria...? Definiteness only, not semidef.

First, in order for the following to work,
the leftmost $m \times m$ minor of the full-rank
 $m \times n$ matrix \vec{B} must be nonzero.

If not: re-enumerate the variables, redefine
 \vec{A} and \vec{B} .

Form the bordered Hessian

$$\begin{pmatrix} \vec{0}_{m \times m} & \vec{B} \\ \vec{B}^T & \vec{A} \end{pmatrix}$$

Let b_r = the leading
principal $(m+r) \times (m+r)$
minor - covering "down
to variable r "

Then:

\mathcal{Q} pos. def. subject to $\vec{B}\vec{x} = \vec{0}$



$(-1)^m b_r > 0$ for all $r = m+1, \dots, n$



\mathcal{Q} neg. def. subject to $\vec{B}\vec{x} = \vec{0}$



$(-1)^r b_r > 0$ for all $r \geq m+1$.

Example:

\vec{A} indef.

$$\vec{A} = \begin{pmatrix} 1 & 0 & 0 & -1 \\ 0 & 1 & 2 & 0 \\ 0 & 2 & -1 & 0 \\ -1 & 0 & 0 & 1 \end{pmatrix}, \quad \vec{B} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 2 & 0 \\ 1 & 0 & -9 & 1 \end{pmatrix}$$

Let's be lazy. First row of \vec{B} says $x_1 = 0$.

Just put $x_1 = 0$ to get the following in $\vec{y} = \begin{pmatrix} x_2 \\ x_3 \\ x_4 \end{pmatrix}$:

$$\vec{A}_{-1} = \begin{pmatrix} 1 & 2 & 0 \\ 2 & -1 & 0 \\ 0 & 0 & 1 \end{pmatrix}, \quad \vec{B}_{-1} = \begin{pmatrix} 1 & 1 & 0 \\ 0 & -9 & 1 \end{pmatrix}$$

nonzero minor

Form

$$\begin{pmatrix} 0 & 0 & \vec{B} \\ \vec{B}^T & \vec{A} \end{pmatrix} = \begin{pmatrix} 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & -9 & 1 \\ 1 & 0 & 1 & 2 & 0 \\ 1 & -9 & 2 & -1 & 0 \\ 0 & 1 & 0 & 0 & 1 \end{pmatrix}$$

I shall never check \det , remember?

2 constraints, b_2 is 3×3 , shall check

b_{m+1} and up, b_{m+1} is 5×5 .

Cofactor exp: $\begin{vmatrix} 0 & 0 & 1 & 1 \\ 0 & 0 & 0 & -9 \\ 1 & 0 & 1 & 2 \\ 1 & -9 & 2 & -1 \end{vmatrix} - \begin{vmatrix} 0 & 1 & 1 & 0 \\ 0 & 0 & -9 & 1 \\ 1 & 1 & 2 & 0 \\ 1 & -2 & -1 & 0 \end{vmatrix}$

$$= -81 \begin{vmatrix} 1 & 1 \\ 1 & 1 \end{vmatrix} - \begin{vmatrix} 0 & 1 & 1 \\ 1 & 1 & 2 \\ 1 & 2 & -1 \end{vmatrix} = 81 - 4$$

$m = 2$ now, so $(-1)^m b_r > 0, r = 3 \dots 3$. pos. def s.t the constraint